

Student Notes Nov 16

swp293, sm8107

November 16, 2020

1 Lecture Video

1.1 Conditional Distribution: Discrete Case

Conditional Distributions

Goal: Given X, Y random variables with joint pdf $f(x, y)$ and marginal pdfs: f_x, f_y , what is the pdf of the random variables :

$$z = X|Y = y$$

Recall for events:

$$P[E|F] = P[EF]/P[F]$$

Define the **conditional probability mass function** so that:

$$P_{X|Y}(x|y) = P[X = x|Y = y] = \frac{P[X = x, Y = y]}{P[Y = y]} = \frac{p(x, y)}{P_Y(y)}$$

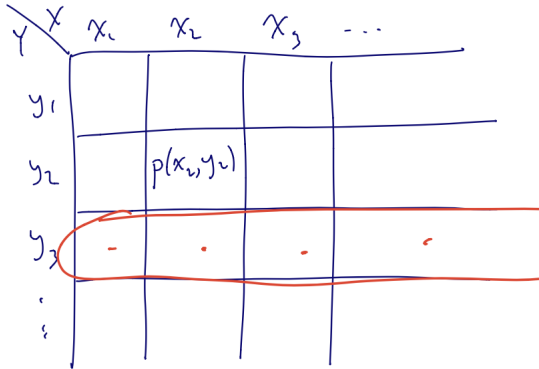
$p(x, y)$: joint mass function

$P_Y(y)$: marginal mass function (Assuming $P_Y(y) > 0$)

Conditional distribution function is then:

$$F_{X|Y}(x|y) = P[X \leq x|Y = y] = \sum_{a \leq x} P[X = a|Y = y]$$

Pictorially, imagine there is a grid of probabilities:



Fix $Y = y_3$, scale this row by $\frac{1}{P_Y(y_3)}$ so that $\sum P_{X|Y}(x_i|y_3)$

If X and Y are **independent random variables**, then:

$$p(x, y) = P_X(x)P_Y(y)$$

$$\rightarrow P_{X|Y}(x|y) = \frac{P(x, y)}{P_Y(y)} = \frac{P_X(x)P_Y(y)}{P_Y(y)} = P_X(x)$$

1.2 Conditional Distribution: Continuous Case

If X and Y have a joint probability density function $f(x, y)$, then the conditional probability density function of X given that $Y = y$ is defined, for all values of y such that $f_Y(y) > 0$, by:

$$f_{X|Y}(x|y) = \frac{f(x, y)}{f_Y(y)}$$

To motivate this definition, multiply the left-hand side by dx and the right-hand side by $(dx dy)/dy$ to obtain

$$\begin{aligned} f_{X|Y}(x|y) &= \frac{f(x, y)dx dy}{f_Y(y)dy} \\ &\cong \frac{P\{x \leq X \leq x + dx, y \leq Y \leq y + dy\}}{P\{y \leq Y \leq y + dy\}} \\ &= P\{x \leq X \leq x + dx, y \leq Y \leq y + dy\} \end{aligned}$$

The use of conditional densities allows us to define conditional probabilities of events associated with one random variable when we are given the value of

a second random variable. That is, if X and Y are jointly continuous, then, for any set A,

$$f_{X|Y}(x|y) = \frac{f(x, y)dx dy}{f_Y(y)dy}$$

The use of conditional densities allows us to define conditional probabilities of events associated with one random variable when we are given the value of a second random variable. That is, if X and Y are jointly continuous, then, for any set A,

$$P(X \in A|Y = y) = \int_A f_{X|Y}(x|y)dx$$

Likewise,

$$P(X \in A|Y \in B) = \int_A \int_B f_{X|Y}(x|y)dx dy$$

For A in range $(-\infty, x)$, the distribution function is :

$$\begin{aligned} f_{X|Y}(x|y) &= P[X \leq x|Y = y] \\ &= \int_{-\infty}^x f_{X|Y}(u|y)du \end{aligned}$$

If X and Y are independent, then:

$$f_{X|Y}(x|y) = f_X(x)$$

2 In Class Examples

2.1 Discrete Conditional Distribution

$$f(x, y) = \begin{cases} \frac{e^{-x}}{y} e^{-y}, & \text{if } x, y > 1 \\ 0, & \text{otherwise} \end{cases}$$

What is $P[X > 1|Y = y]$?

$$f_{X|Y}(x|y) = \frac{f(x, y)}{f_Y(y)}$$

$$\begin{aligned}
f_Y(y) &= \int_0^\infty f(x,y)dx = \int_0^\infty \frac{e^{-\frac{x}{y}} e^{-y}}{y} dx \\
&= \frac{e^{-y}}{y} \int_0^\infty e^{-\frac{x}{y}} dx \\
&= \left(\frac{e^{-y}}{y} \right) - ye^{\frac{x}{y}} \Big|_0^\infty \\
&= \frac{e^{-y}}{y} (0 + ye^0) \\
&= e^{-y}
\end{aligned} \tag{1}$$

$$\begin{aligned}
f_{X|Y}(x|y) &= \frac{f(x,y)}{f_Y(y)} \\
&= \frac{e^{-\frac{x}{y}} e^{-y}/y}{e^{-y}} \\
&= \frac{e^{-\frac{x}{y}}}{y}
\end{aligned} \tag{2}$$

Then,

$$\begin{aligned}
P[X > 1|Y = y] &= \int_1^\infty f_{X|Y}(x|y)dx = \int_1^\infty \frac{e^{-\frac{x}{y}}}{y} dx \\
&= \left(\frac{1}{y} \right) - ye^{-\frac{x}{y}} \Big|_1^\infty \\
&= \left(\frac{1}{y} \right) (0 + ye^{-\frac{1}{y}}) \\
&= e^{-\frac{1}{y}}
\end{aligned} \tag{3}$$

Example 4B (pg. 264):

If X and Y are independent Poisson variables with respective parameters λ_1 and λ_2 , calculate the conditional distribution of X given that $X+Y = n$.

We calculate the conditional probability mass function of X given that $X+Y=n$ as follows:

$$P[X = k|X + Y = n] = \frac{P[X = k, X + Y = n]}{P[X + Y = n]} = \frac{P[X = k, Y = n - k]}{P[X + Y = n]} = \frac{P[X = k]P[Y = n - k]}{P[X + Y = n]}$$

The last equality follows from the assumed independence of X and Y. Recalling that X+Y has a Poisson distribution with parameter $\lambda_1 + \lambda_2$, we see that it equals:

$$\begin{aligned}
 P[X = k | X + Y = n] &= \left(\frac{e^{\lambda_1} \lambda_1^k}{k!} \right) \left(\frac{e^{-\lambda_2} \lambda_2^{n-k}}{(n-k)!} \right) \left[\frac{e^{-(\lambda_1 + \lambda_2)} (\lambda_1 + \lambda_2)^n}{n!} \right]^{-1} \\
 &= \left(\frac{n!}{(n-k)!k!} \right) \left(\frac{\lambda_1^k \lambda_2^{n-k}}{(\lambda_1 + \lambda_2)^n} \right) \\
 &= \binom{n}{k} \left(\frac{\lambda_1}{\lambda_1 + \lambda_2} \right)^k \left(\frac{\lambda_2}{\lambda_1 + \lambda_2} \right)^{n-k}
 \end{aligned} \tag{4}$$

The conditional distribution of X given that X+Y=n is the binomial distribution with parameters n and $\frac{\lambda_1}{\lambda_1 + \lambda_2}$.

2.2 Continuous Conditional Distribution

5a (page 266):

The joint density of X and Y is given by:

$$f(x, y) = \begin{cases} 12x(2 - x - y), & 0 < x < 1, 0 < y < 1 \\ 0, & \text{otherwise} \end{cases} \tag{5}$$

Question: Compute the conditional density of X given that Y=y, given that $0 < y < 1$.

Solution: For $0 < x < 1, 0 < y < 1$, we have:

$$\begin{aligned}
 f_{X|Y}(x|y) &= \frac{f(x, y)}{f_Y(y)} \\
 &= \frac{f(x, y)}{\int_{-\infty}^{+\infty} f(x, y) dx} \\
 &= \frac{x(2 - x - y)}{\int_0^1 x(2 - x - y) dx} \\
 &= \frac{x(2 - x - y)}{2/3 - y/2} \\
 &= \frac{6x(2 - x - y)}{4 - 3y}
 \end{aligned}$$

5b (page 267): Suppose that the joint density of X and Y is given by:

$$f(x, y) = \begin{cases} \frac{e^{-x/y} e^{-y}}{y}, & 0 < x < \infty, 0 < y < \infty \\ 0, & \text{otherwise} \end{cases} \tag{6}$$

Find $P[X > 1|Y = y]$.

Solution:

$$\begin{aligned} f_{X|Y}(x|y) &= \frac{f(x, y)}{f_Y(y)} \\ &= \frac{e^{-x/y} e^{-y}/y}{e^{-y} \int_0^{\infty} (1/y) e^{-x/y} dx} \\ &= \frac{e^{-x/y}}{y} \end{aligned}$$

Hence,

$$\begin{aligned} P(X > 1|Y = y) &= \int_1^{\infty} e^{-x/y}/y dx \\ &= e^{-1/y} \end{aligned}$$

When X and Y are independent,

$$f_{X|Y}(x|y) = \frac{f(x, y)}{f_Y(y)} = \frac{f_X(x)f_Y(y)}{f_Y(y)} = f_X(x)$$

Suppose that X is a continuous random variable having probability density function f and N is a discrete random variable, and consider the conditional distribution of X given that N = n. Then:

$$\frac{P(x < X < x + dx|N = n)}{dx} = \frac{P(N = n|x < X < x + dx)P(x < X < x + dx)}{P(N = n)dx}$$

Letting dx approach 0, then:

$$\lim_{dx \rightarrow 0} \frac{P(x < X < x + dx|N = n)}{dx} = \frac{P(N = n|X = x)f(x)}{N = n}$$

Therefore,

$$f_{X|N}(x|n) = \frac{P(N = n|X = x)f(x)}{N = n}$$